Speech Systems - Course Introduction

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- Economical
 - Inexpensive transmission and reception of information
 - Voice communication is the reason behind success of mobile phones

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 - Background Information Acoustic environment around the speaker

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 - Coding and transmission of speech signals
- Signal processing
 - Signal detection, enhancement and feature extraction
- AI & Machine Learning
 - Data-driven models for information extraction
 - Speech recognition, speech synthesis, speaker recognition, language recognition, etc.

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 - Study of speech sounds through analysis of waveform and spectrum
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- Building speech systems
 - Probability density estimation Gaussian mixture models
 - Extracting utterance-level representation from speech signal
 - Speaker recognition, language recognition and emotion recognition

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- Deep acoustic modeling
 - Discriminative models for likelihood estimation using DNNs (pretend)
 - End-to-end models for ASR (waveform to character)



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- Speech synthesis
 - Estimate joint density from marginals.
 - III posed problem Difficult to infer dependencies
 - Draw a sample function from the estimated joint density

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Suggested Prerequisites

- Basics of signal processing
- Probability and random processes
- Foundations of machine learning
 - Supervised learning, regression, classification
- Neural networks
 - Understanding of forward and back propagation equations
 - Basic exposure to deep learning
- Proficiency in Python

Evaluation Criteria

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- Programming assignments have to be done in Python.
- Project theme has to be finalized by the end of the first segment
- Handwritten summary of lecture has to be submitted on the same day
- Delayed submissions will be penalized

Thank You!